



An algorithmic framework for developing saturation rules in reduced core transport models

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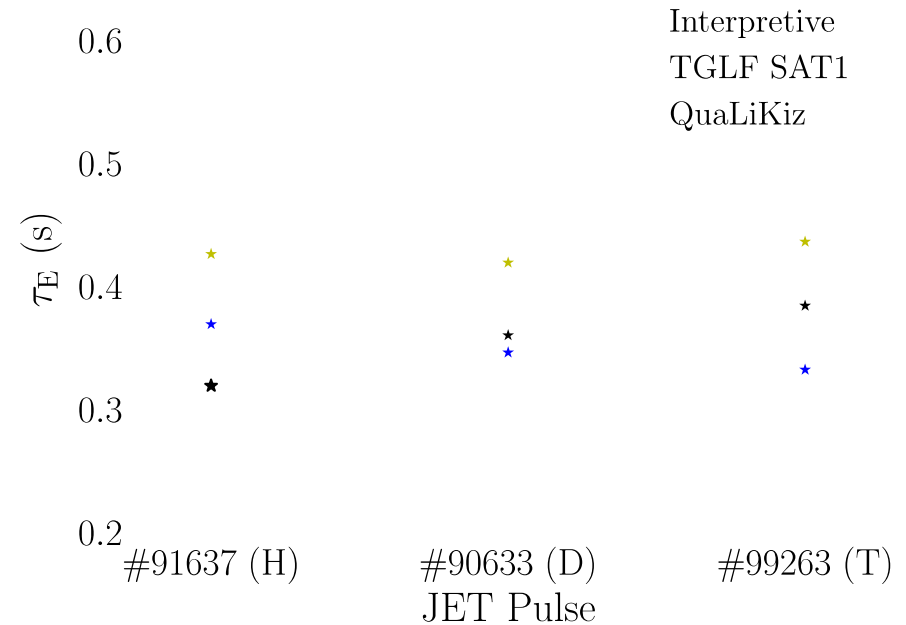
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Introduction

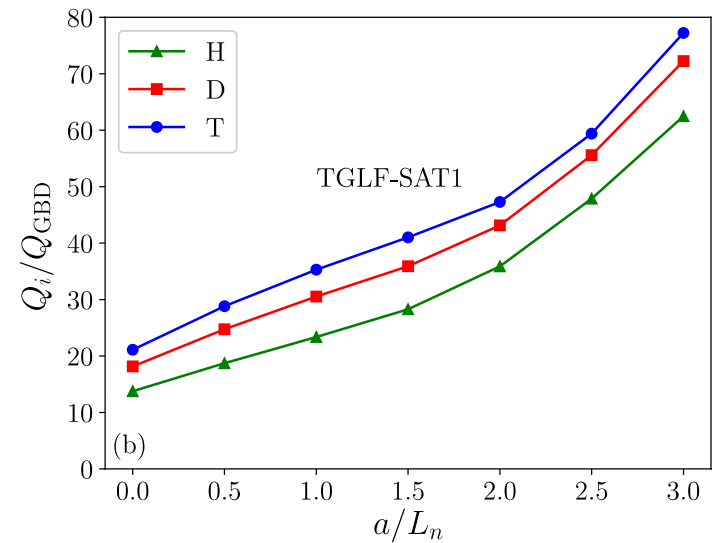
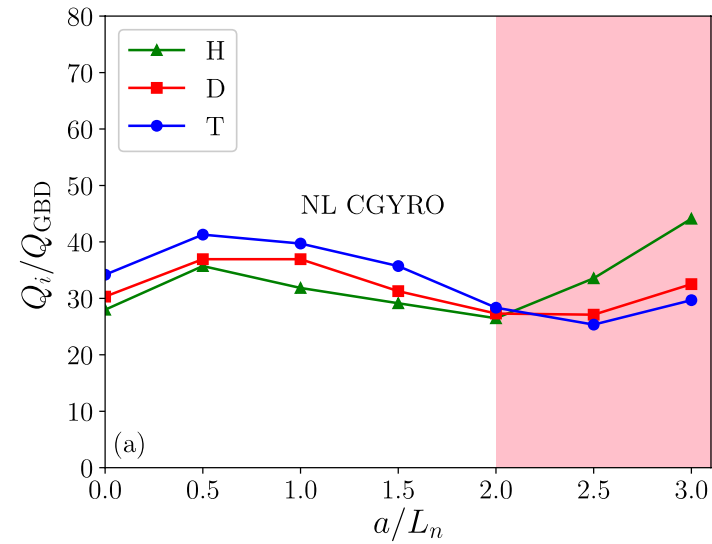
- Reduced turbulence models are needed for integrated modelling due to relatively low computational expense
- Quasilinear (QL) models (TGLF, QuaLiKiz) approximate fluxes using *simplified* linear physics and a **saturation rule**
- Saturation rules built from theory and fits to NL GK simulations
⇒ can extrapolate poorly to new parameter space



Introduction

- Typically tuned on **large aspect-ratio, electrostatic, deuterium** plasmas
- We require *validated* transport models for current and future experiments, particularly in areas of e.g:

- Fast ions
- Mixed plasmas
- Plasma shaping
- Electromagnetic turbulence/ High β



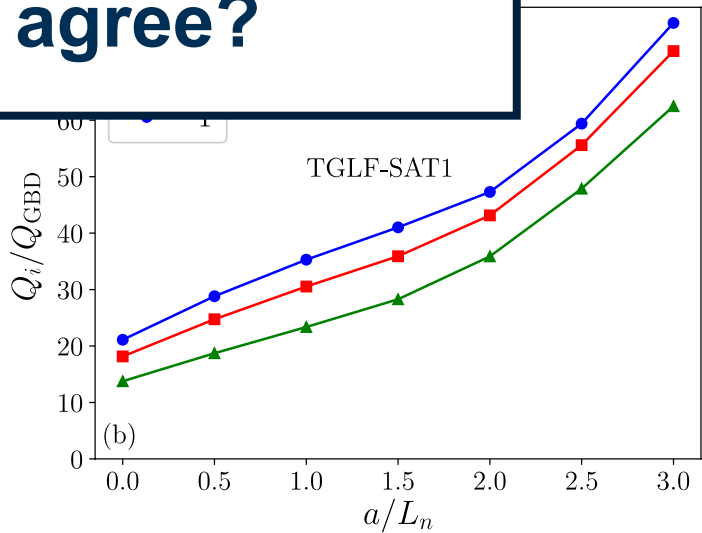
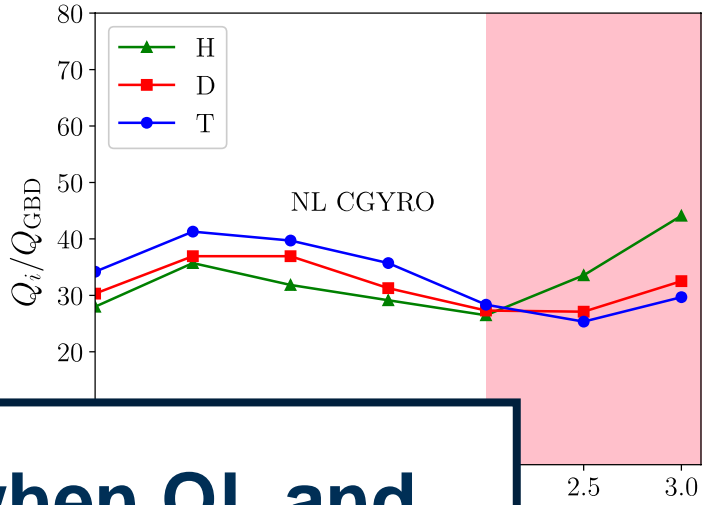
- Test models via comparison with standalone NL GK simulations
- In this talk, focus on the development of the new saturation rule SAT3 from discrepancies in **isotope scaling**

Introduction

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- We require *validated* transport models for current and future experiments, particularly in areas of e.g:

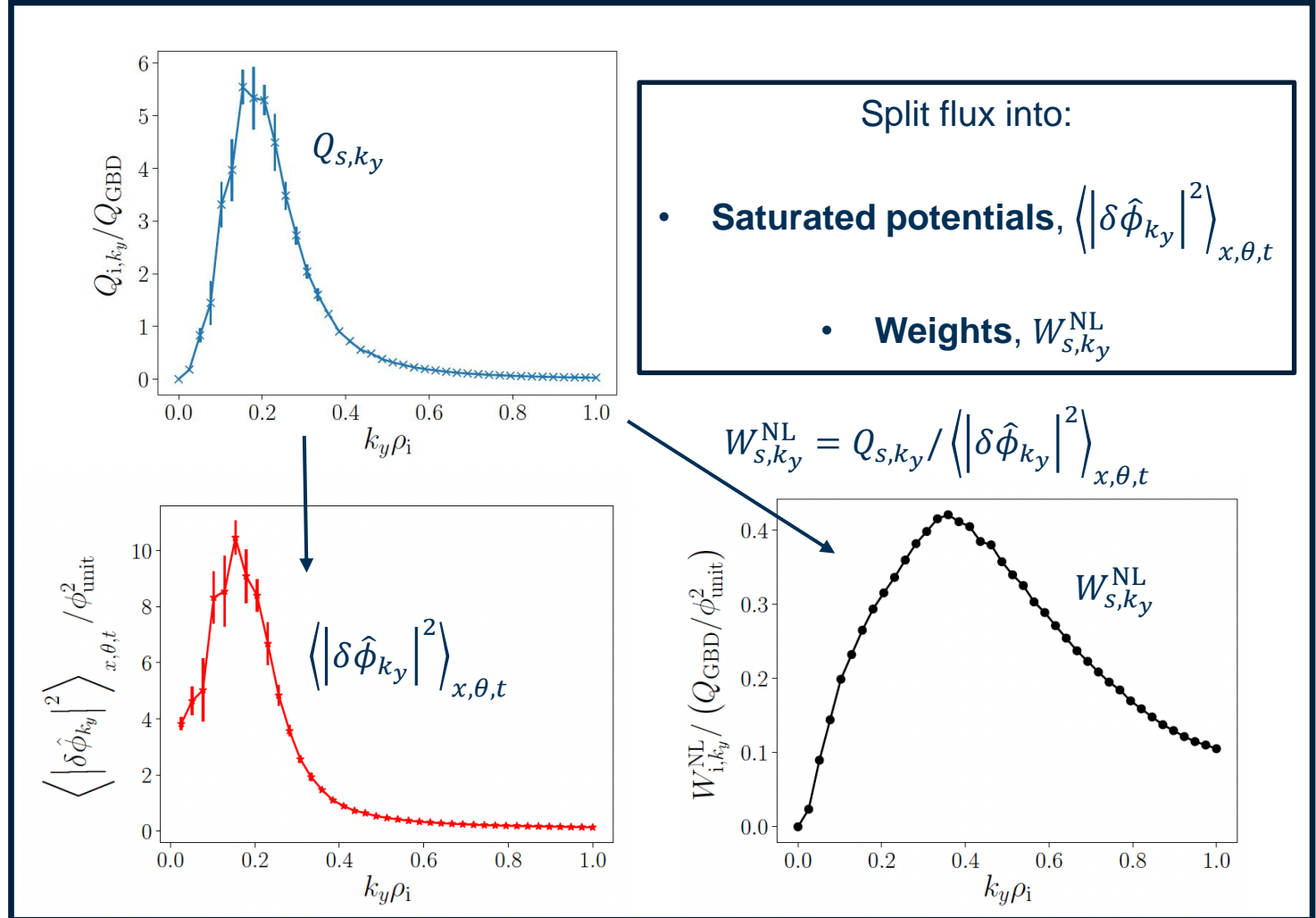
- Fast ion
- Plasma shaping

What do we do when QL and NL GK don't agree?



Turbulent flux theory

$$Q_s = \sum_{k_y > 0} Q_{s,k_y}$$



Turbulent flux theory

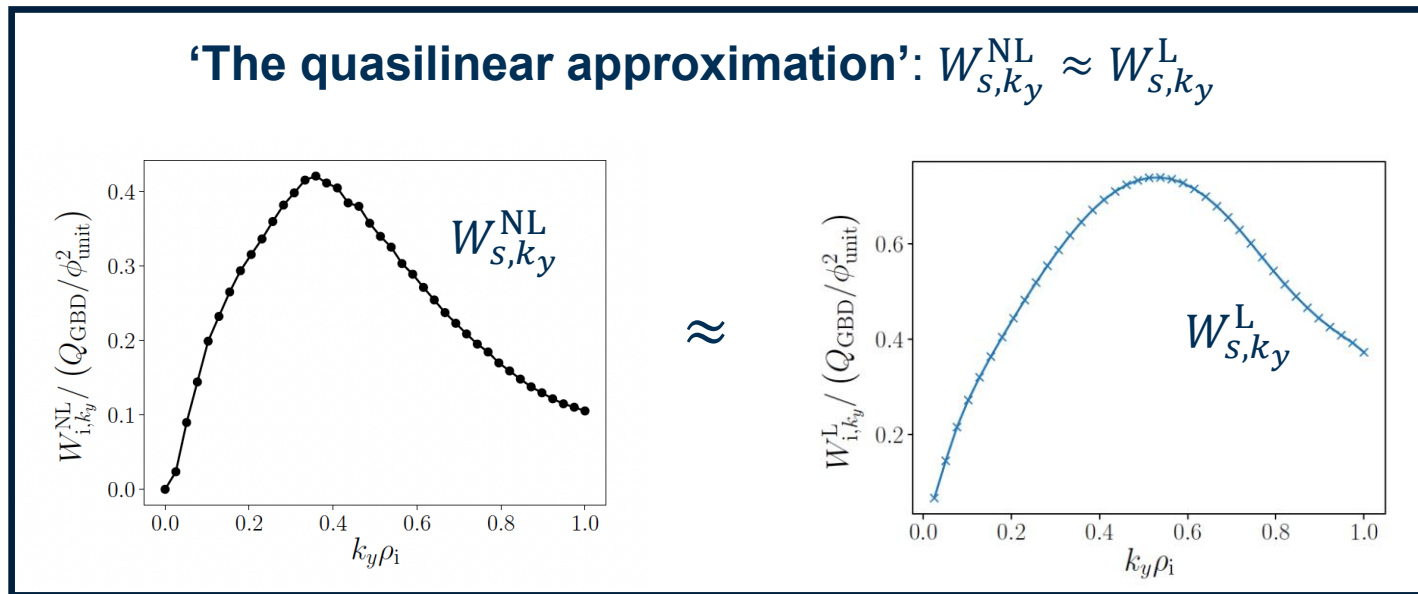
$$Q_s = \sum_{k_y > 0} Q_{s,k_y}$$

↓

$$= \sum_{k_y > 0} W_{s,k_y}^{\text{NL}} \langle |\delta \hat{\phi}_{k_y}|^2 \rangle_{x,\theta,t}$$

Turbulent flux theory: The quasilinear approximation

- W_{S,k_y}^{NL} calculated from **saturated** turbulence in **nonlinear** simulation
- Can also calculate the phase difference in a **linear** simulation, W_{S,k_y}^L



- Quantify departure from perfect agreement via

$$W_{S,k_y}^{NL} = \Lambda_{S,k_y} W_{S,k_y}^L$$

Turbulent flux theory

$$Q_s = \sum_{k_y > 0} Q_{s,k_y}$$

$$= \sum_{k_y > 0} W_{s,k_y}^{\text{NL}} \left\langle \left| \delta \hat{\phi}_{k_y} \right|^2 \right\rangle_{x,\theta,t}$$



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Quasilinear models:

Turbulent flux theory

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Quasilinear models:

Λ_{s,k_y} : Assume $\approx \text{const.}$
(QL approximation)

Turbulent flux theory

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Λ_{s,k_y} : Assume $\approx \text{const.}$
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W_{s,k_y}^{L} : Approximate using
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Turbulent flux theory

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Quasilinear models:

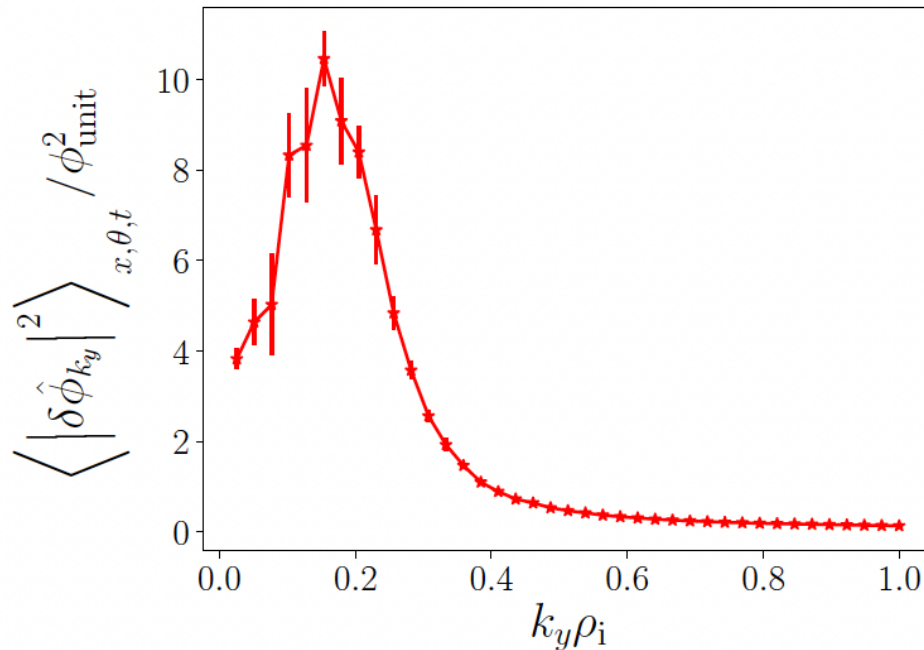
Λ_{s,k_y} : Assume $\approx \text{const.}$
(QL approximation)

W_{s,k_y}^{L} : Approximate using
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$\left\langle \left| \delta \hat{\phi}_{k_y} \right|^2 \right\rangle_{x,\theta,t}$: Approximate using
saturation rule

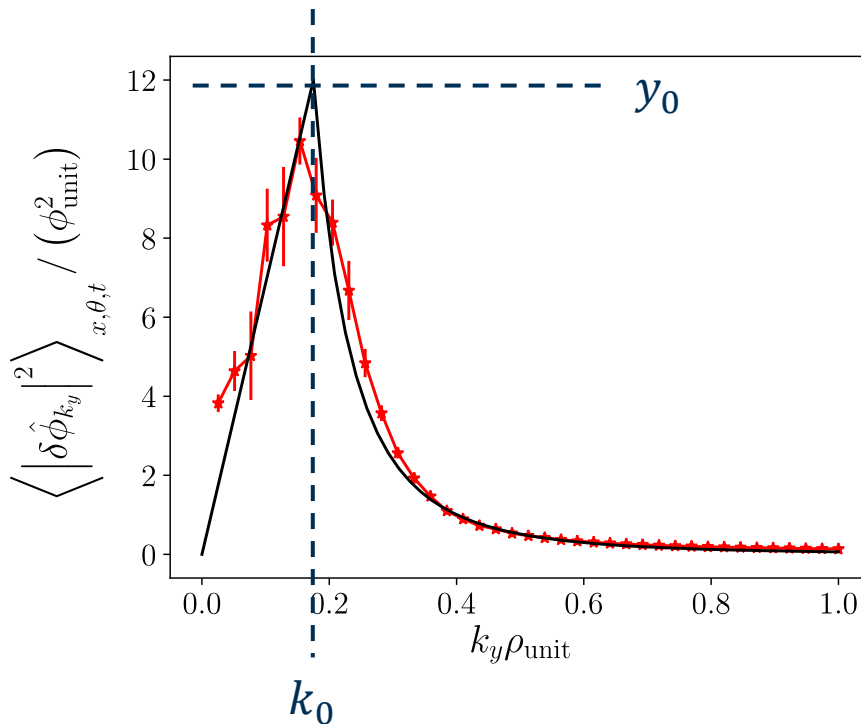
Saturation Rules

- Saturation **inherently nonlinear** process \Rightarrow cannot invoke linear gyrokinetics like with the weights
- Saturation rules guided by theory and fits to NL GK data to predict potential spectra



Saturation Rules: QLK example

$$y(k_y) = \begin{cases} y_0 \left(\frac{k_y}{k_0} \right) & 0 < k_y \leq k_0 \\ y_0 \left(\frac{k_y}{k_0} \right)^{-3} & k_0 < k_y < \infty \end{cases}$$



$$y(k_y) = y_0 S(k_y)$$

$S(k_y)$: 'Spectral shape'

y_0 : 'Saturation level'

Still need to model y_0, k_0 !

Saturation rules: linear modelling

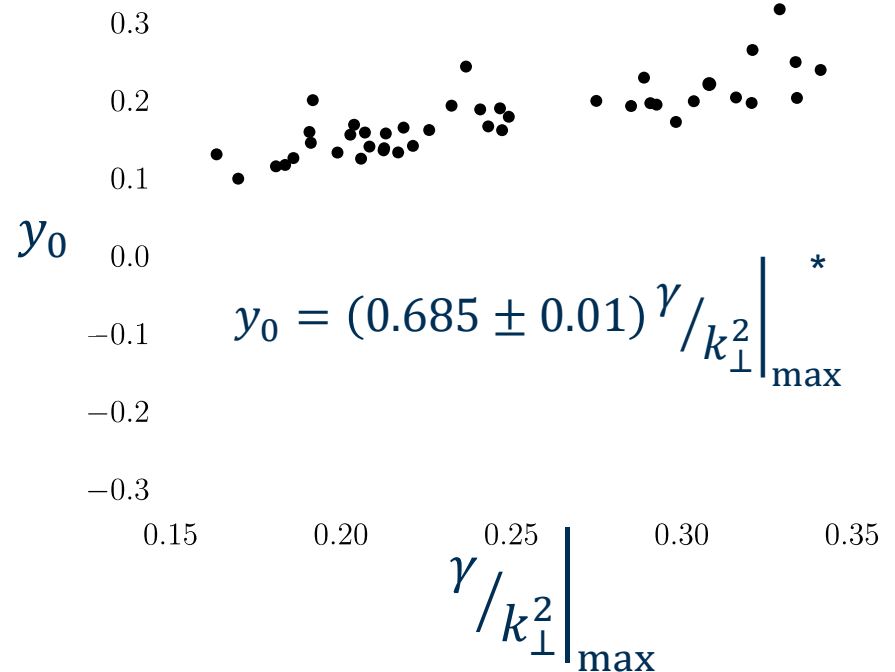
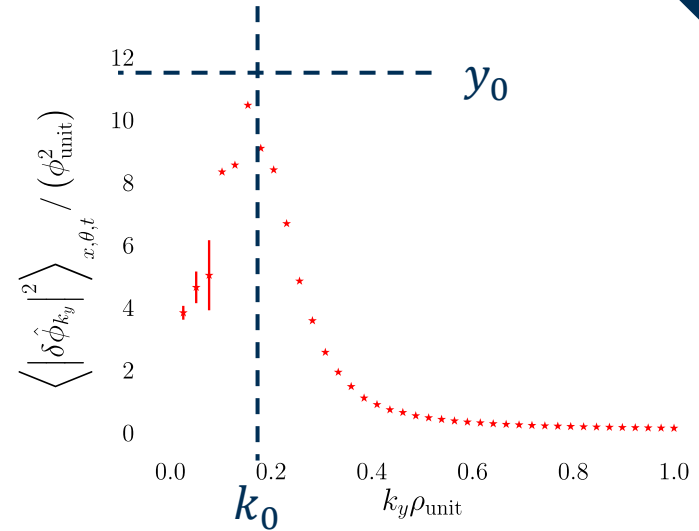
- Parameters vary case-by-case \Rightarrow require 'linear model' for each one
- Use physics arguments to relate parameters to linear properties, e.g.

Mixing length rule

- Consider transport as *diffusive* process
- Argue that potential peak varies with diffusion coefficient

$$y_0 \sim D_{\perp} \sim \frac{\Delta l^2}{\Delta t} \propto \frac{\gamma}{k_{\perp}^2} \Big|_{\max}$$

- Recommended to fit using linear gyrokinetics, **not** simplified linear solvers

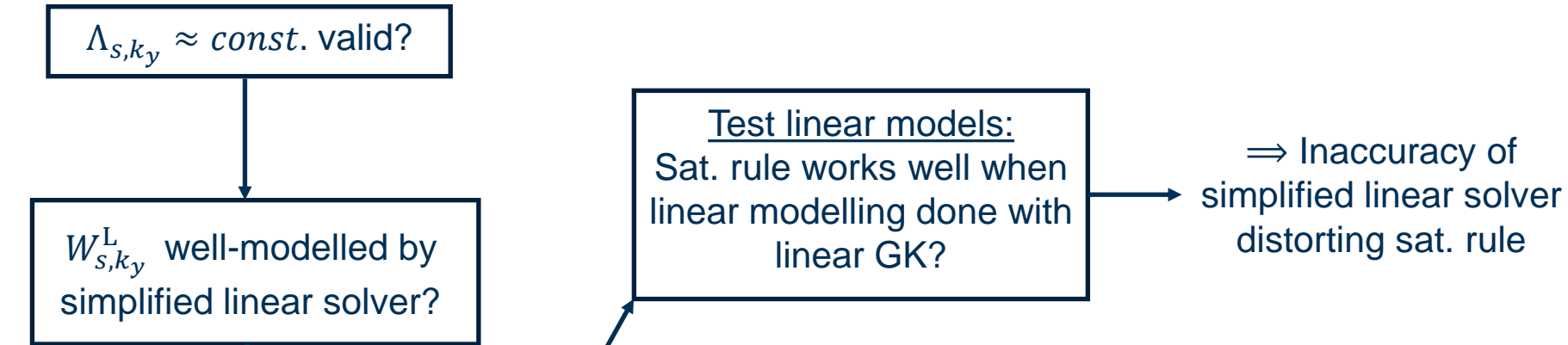


*Note: Toy data for illustrative purposes

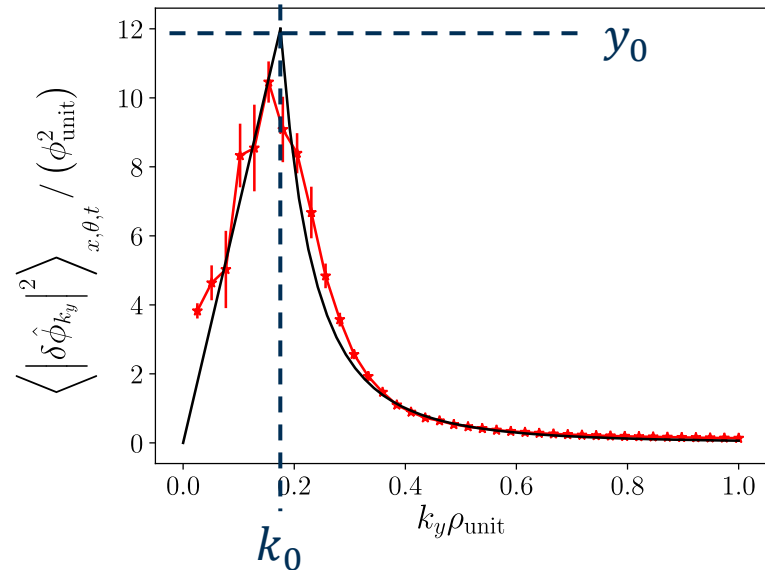
Discrepancy algorithm

$$Q_s = \sum_{k_y > 0} \Lambda_{s,k_y} W_{s,k_y}^L \langle |\delta \hat{\phi}_{k_y}|^2 \rangle_{x,\theta,t}$$

Consider each aspect in turn, develop where necessary:



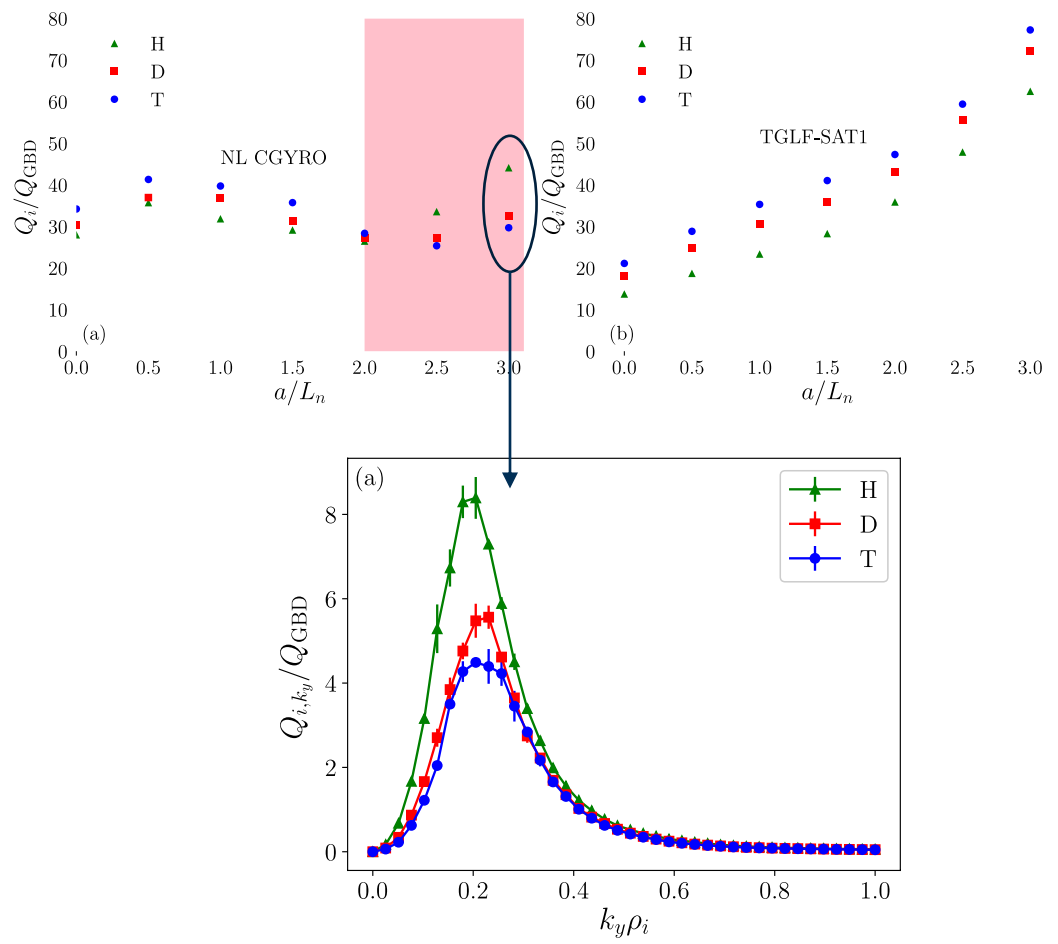
Test spectral shape:
Sat. rule works well when parameters (e.g. γ_0, k_0) fitted to NL GK data?



Addressing isotope scaling in TGLF

$$Q_s = \sum_{k_y > 0} \Lambda_{s,k_y} W_{s,k_y}^L \left\langle \left| \delta \hat{\phi}_{k_y} \right|^2 \right\rangle_{x,\theta,t}$$

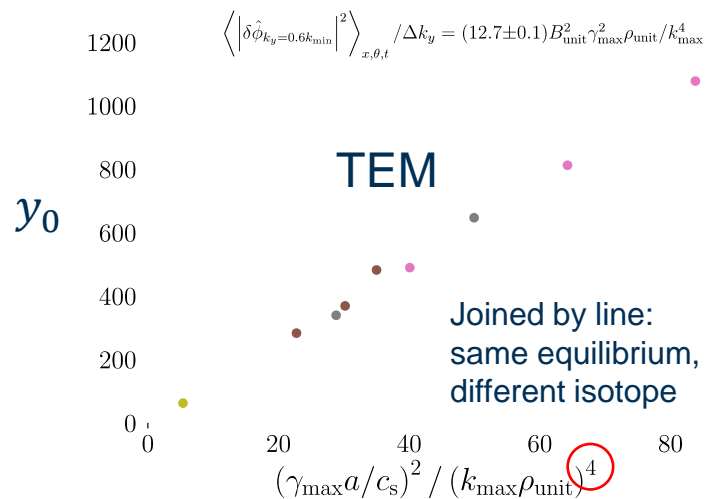
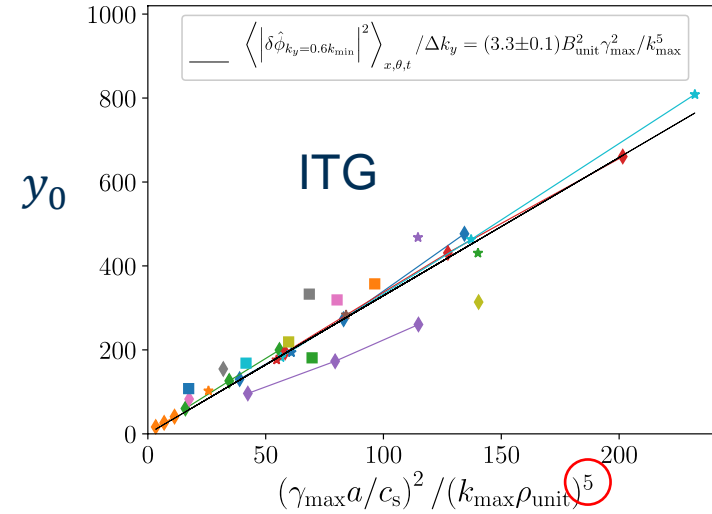
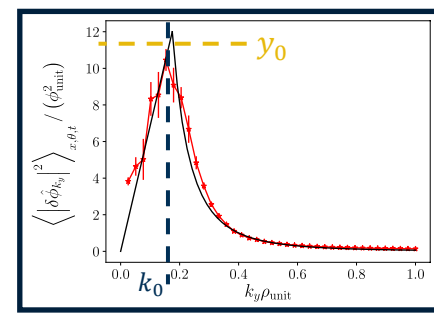
- Discrepancy between **isotope scaling** of fluxes in TEM-dominant regime
- \Rightarrow Build NL GK database of ~ 50 simulations, expanding on previous to include different isotopes (H, D, T)
- Begin discrepancy algorithm, find isotope scaling to originate in region of *low* k_y
- Found $\Lambda_{s,k_y} \approx const.$ and modelling of W_{s,k_y}^L to be well-satisfied \Rightarrow Turn to the saturation rule!



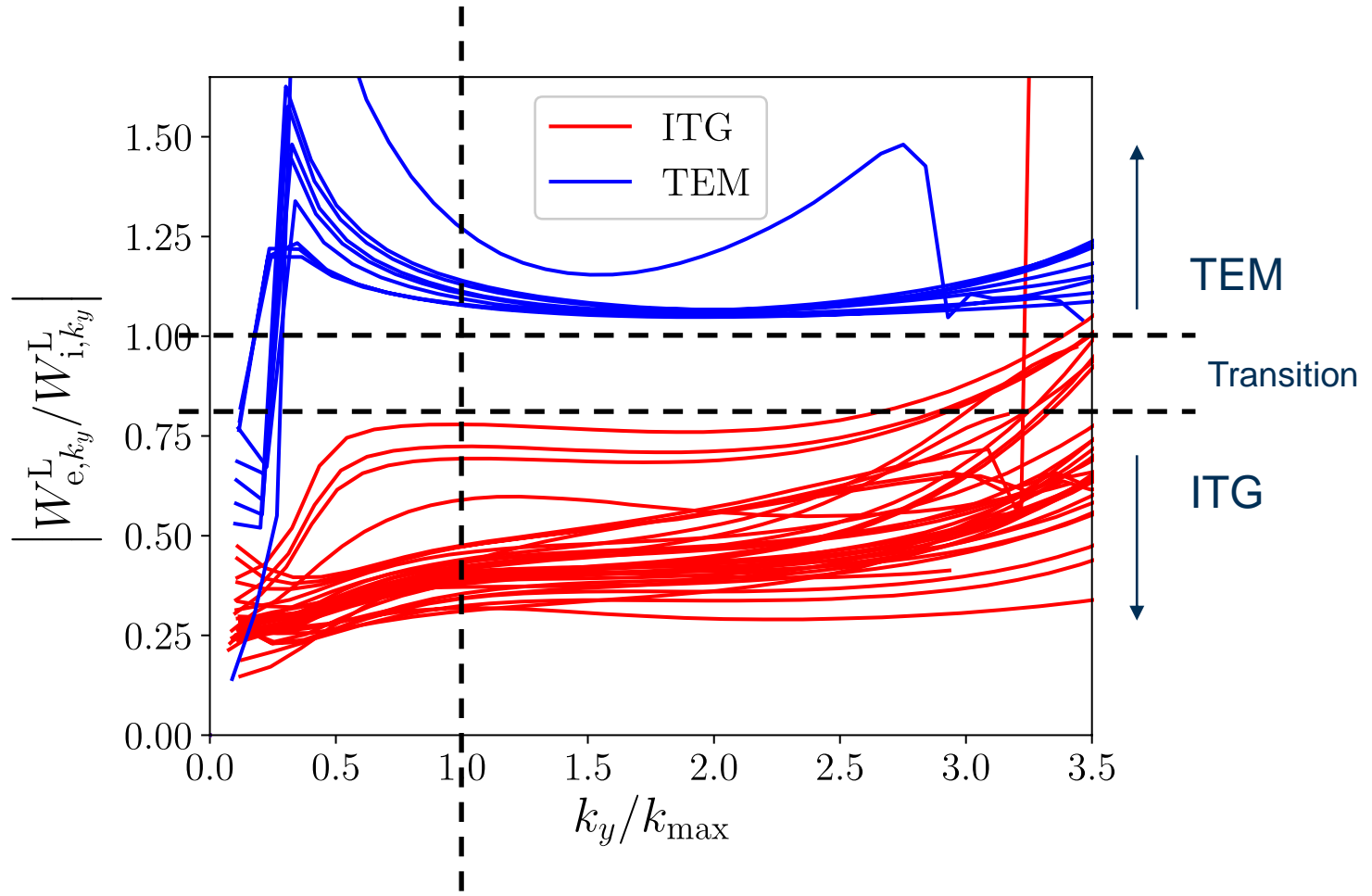
Addressing isotope scaling in TGLF: SAT3

$$Q_s = \sum_{k_y > 0} \Lambda_{s,k_y} W_{s,k_y}^L \left\langle \left| \delta \hat{\phi}_{k_y} \right|^2 \right\rangle_{x,\theta,t}$$

- Found dominant cause of isotope scaling to be linear modelling of the saturation level, y_0
- Find **different** saturation levels depending on if turbulence is ITG or TEM-dominated
- SAT3 model allows for **transitioning** between saturation levels **depending** on dominant mode type

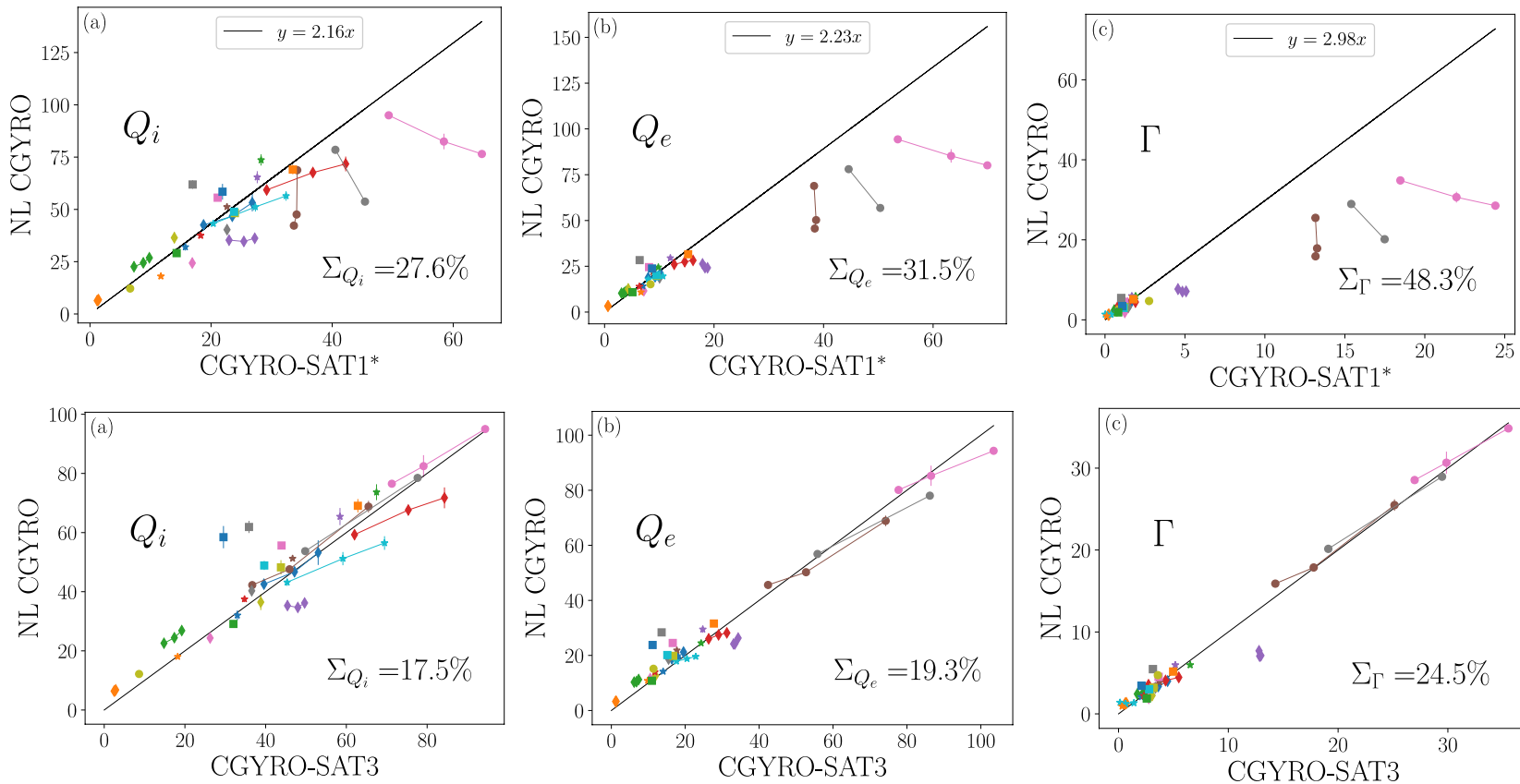


SAT3 mode identification



Addressing isotope scaling in TGLF: SAT3

- These improvements allow us to capture the isotope scaling of TEM turbulence
- Constitutes **extension** of model validity while still performing well in established parameter spaces
- SAT3 available for use on GACODE master branch



Summary

- Quasilinear transport models can perform less-well outside of their tuned parameter space
- Presented an algorithm to address discrepancies, focusing on the separation of each contribution to isolate the root cause
- Algorithm was applied in the development of SAT3, allowing us to correctly model the isotope scaling of TEM turbulence in QL models
- Integrated modelling validation efforts currently being performed

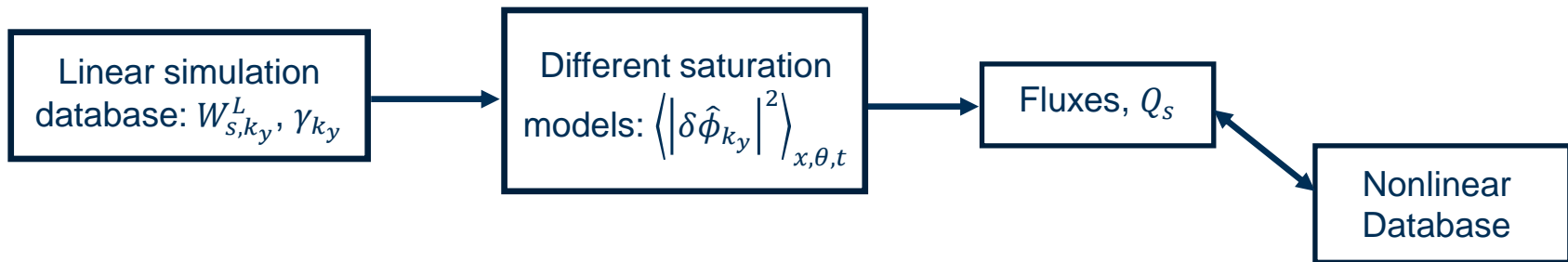
Discussion: future

1. Electromagnetic fluxes:

- 5 Λ_{s,k_y} -like quantities, 5 W_{s,k_y}^L -like quantities, 1 saturation rule
- \Rightarrow Generalise first two steps of the algorithm

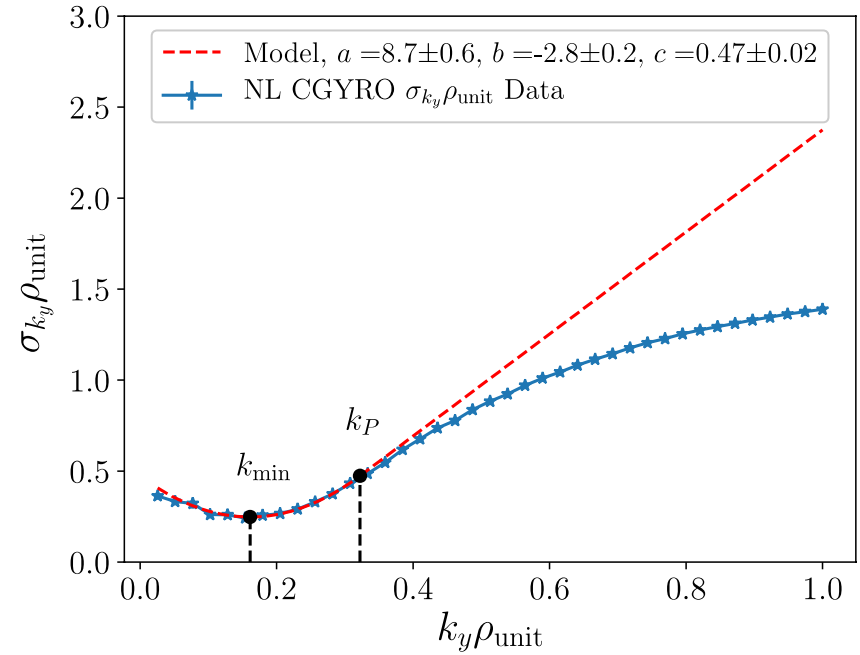
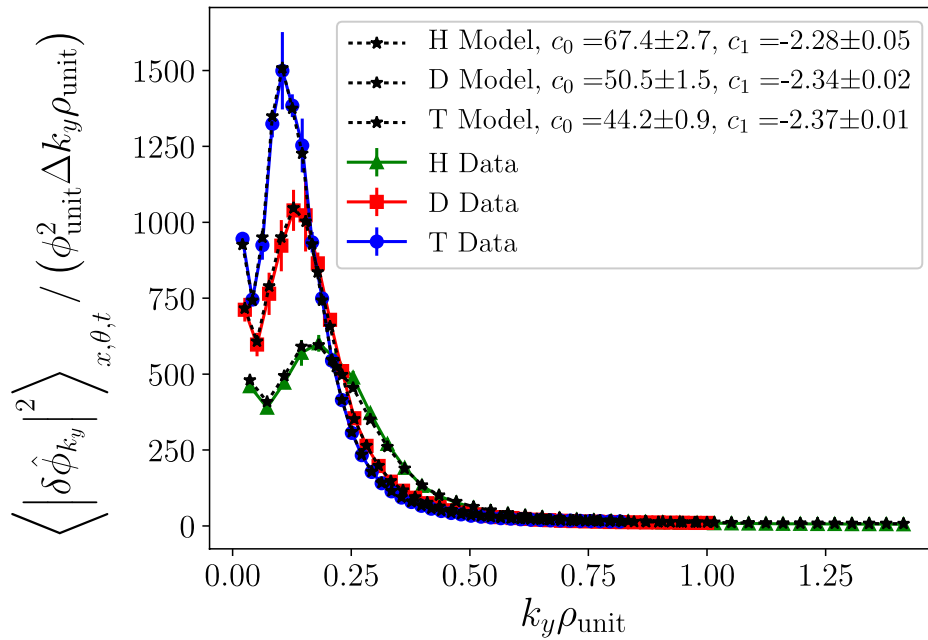
$$\begin{aligned}
 Q_s &= \sum_{k_y > 0} Q_{s,k_y}^{\delta\phi} + Q_{s,k_y}^{\delta A_{||}} + Q_{s,k_y}^{\delta B_{||}} \\
 &= \sum_{k_y > 0} \left[\frac{Q_{s,k_y}^{\delta\phi}}{\langle |\delta\hat{\phi}_{k_y}|^2 \rangle} \langle |\delta\hat{\phi}_{k_y}|^2 \rangle + \frac{Q_{s,k_y}^{\delta A_{||}}}{\langle |\delta\hat{A}_{||,k_y}|^2 \rangle} \langle |\delta\hat{A}_{||,k_y}|^2 \rangle + \frac{Q_{s,k_y}^{\delta B_{||}}}{\langle |\delta\hat{B}_{||,k_y}|^2 \rangle} \langle |\delta\hat{B}_{||,k_y}|^2 \rangle \right] \\
 &= \sum_{k_y > 0} \left(\left[\frac{Q_{s,k_y}^{\delta\phi}}{\langle |\delta\hat{\phi}_{k_y}|^2 \rangle} \right] + \left[\frac{\langle |\delta\hat{A}_{||,k_y}|^2 \rangle}{\langle |\delta\hat{\phi}_{k_y}|^2 \rangle} \right] \left[\frac{Q_{s,k_y}^{\delta A_{||}}}{\langle |\delta\hat{A}_{||,k_y}|^2 \rangle} \right] + \left[\frac{\langle |\delta\hat{B}_{||,k_y}|^2 \rangle}{\langle |\delta\hat{\phi}_{k_y}|^2 \rangle} \right] \left[\frac{Q_{s,k_y}^{\delta B_{||}}}{\langle |\delta\hat{B}_{||,k_y}|^2 \rangle} \right] \right) \langle |\delta\hat{\phi}_{k_y}|^2 \rangle
 \end{aligned}$$

2. Necessity of large databases (e.g. GKDB) and community tools for future saturation rule validation



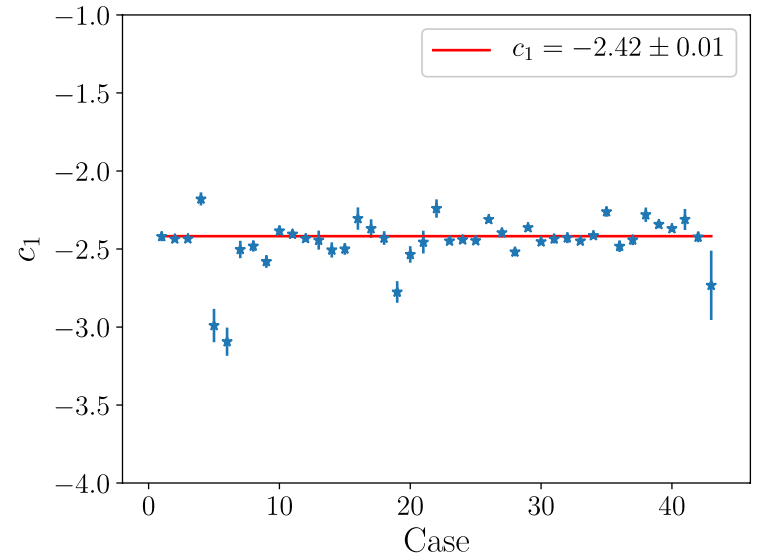
3. Machine learning approach to linear physics: could offer faster and more accurate linear calculations in well-explored regimes

Backup: SAT3 spectral shape



$$\frac{\langle |\hat{\delta\phi}_{k_y}|^2 \rangle_{x,\theta,t}}{\Delta k_y} = c_0 \sigma_{k_y}^{c_1}$$

$$\sigma_{k_y} = \begin{cases} ak_y^2 + bk_y + c & 0 < k_y \leq k_P \\ (2ak_P + b)k_y + c - ak_P^2 & k_P < k_y < \infty \end{cases}$$



Backup: Turbulent flux theory

$$Q_s = 2 \sum_{k_y > 0} \sum_{k_x} \frac{k_y}{B_{\text{ref}}} \left\langle \text{Im} \left[\delta \hat{p}_{s,k_x,k_y}^* \delta \hat{\phi}_{k_x,k_y} \right] \right\rangle_{\theta,t}$$

- Turbulent fluxes arise from interactions between:

- Pressure fluctuations, $\delta \hat{p}_{s,k_x,k_y}^*$
- Potential fluctuations, $\delta \hat{\phi}_{k_x,k_y}$

- For our purposes, sufficient to write

$$Q_s = \sum_{k_y > 0} Q_{s,k_y}$$